



## Statistical Learning 1

Summer semester 2024

Tübingen, 10.06.2024

### Assignment 8

#### Problem 1

To obtain weak universal consistency for the  $k$ -NN estimator, we verify the assumptions in Stone's theorem. In Assignment 7 Problem 1, we have already shown conditions (ii) and (iv)<sub>2</sub> and condition (iii) was proved in the lecture. Show the remaining assumptions under the conditions

$$k_n \uparrow \infty \text{ and } \frac{k_n}{n} \downarrow 0 \text{ for } n \uparrow \infty.$$

#### Problem 2

Show that for the support of  $X$  holds

- $\mathbb{P}[X \in \text{supp}(\mathbb{P}_X)] = 1$
- $\text{supp}(\mathbb{P}_X)$  is closed.

#### Problem 3

Let  $\mathcal{S} \equiv \text{supp}(\mathbb{P}_X)$ . The following estimate is to prove rates for the kernel estimator. Show that a constant  $C_d \equiv C(d, \text{diam}(\mathcal{S})) > 0$  exists, such that

$$\int_{\mathcal{S}} \frac{1}{n \mathbb{P}_X[B(x, h_n)]} \mu[dx] \leq \frac{C_d}{n \cdot h_n^d}.$$

#### Problem 4

For the proof of rates for the kernel estimator, a variance-bias decomposition is used again. Let  $m_n$  be a local averaging estimator and  $m$  the regression function. Show that

$$\begin{aligned} & \mathbb{E}[|m_n(x) - m(x)|^2 | X_1, \dots, X_n] \\ &= \mathbb{E}[|m_n(x) - \mathbb{E}[m_n(x) | X_1, \dots, X_n]|^2 | X_1, \dots, X_n] + \left| \mathbb{E}[m_n(x) | X_1, \dots, X_n] - m(x) \right|^2 \end{aligned}$$

holds for every  $x \in \mathbb{R}^d$ .

**Date of Submission: 17.06.2024 in the mailbox at 12 noon.**